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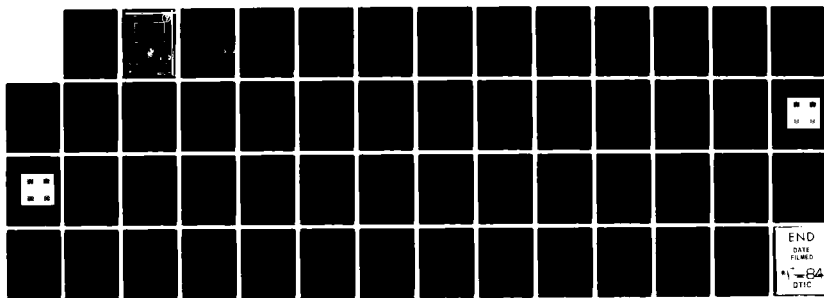
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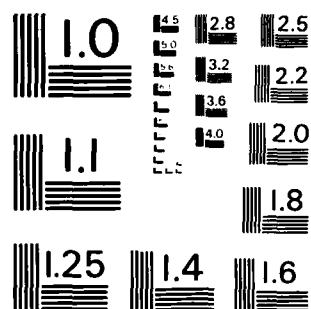
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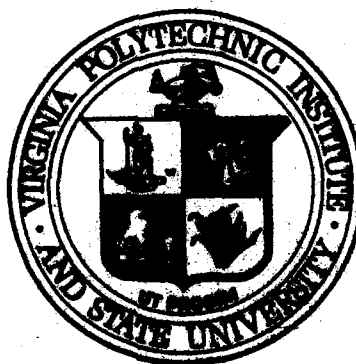
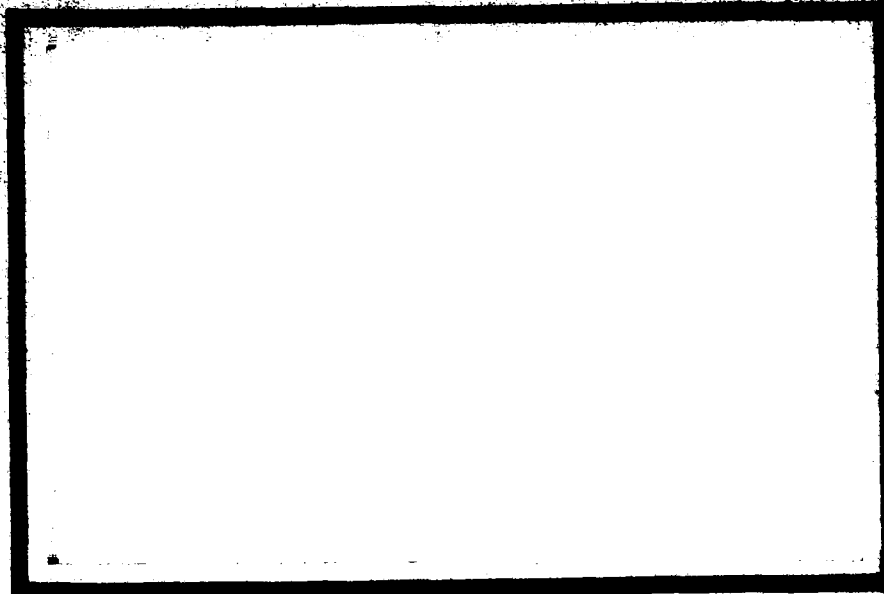




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# Virginia Polytechnic Institute and State University

Electrical Engineering  
BLACKSBURG, VIRGINIA 24061

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FINAL REPORT TO AFOSR

RECURSIVE INTERPOLATION OF SPACE-LIMITED SCENES

A. A. (Louis) Beex

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ITEM #20, CONTINUED: the algorithm iterate to deviate from the noisy measurements, in recognition of the fact that the measurements were noisy, and they should therefore not be imposed as an absolute or hard constraint. The resulting alternating projection algorithm was shown to correspond to a non-expansive operator so that many solutions exist, all of which satisfy the a priori information and constraints.

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FINAL REPORT

RECURSIVE INTERPOLATION OF SPACE-LIMITED SCENES

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## RESEARCH OBJECTIVE

The research performed, and reported on herein, is a continuation of research performed at Rome Air Development Center, Griffiss AFB, during the Summer of 1981 under the USAF-SCEEE Summer Faculty Research Program [1]. This research is therefore directed at the enhancement of resolution in scenes with limited support. The importance to the Air Force lies in the application to space based infrared sensors. In this context we have a field of view limited by a sunshade, operated on by a Fourier transforming lens, and subsequently sampled over a finite support by a detector array.

The idea is to reconstruct the limited field of view scene, most compatible with the given information. This information consists of frequency domain samples, ultimately arrived at by measurements and therefore corrupted by noise, in addition to a priori information. The a priori information yields constraints on solutions to the problem, by such requirements as the nonnegativity of the scene function, the known physical extent of the scene function, and probabilistic characterizations of the corrupting noise.

The objective of the proposed research then, is to evaluate the performance of iterative deconvolution

algorithms, in the context of, and with the constraints for, the space based infrared sensor application. The sensitivity of the solutions with respect to the various constraints will indicate the robustness of the algorithms against assumptions made.

Until now iterative reconstruction algorithms did not incorporate effective ways of dealing with measurement noise. Consequently, the noisy measurements obtained in a practical application, could be incompatible with the a priori constraints. This then leads to searching for a nonexistent solution to the formulated problem. In this stage of the research we concentrate, therefore, on developing modifications of the original algorithm that explicitly allow the use of knowledge about the noise corrupting the measurements. Analysis of the resulting "soft constraint" algorithm shows that the essential convergence properties are preserved. Initial simulation results show the robustness of the new algorithm, when compared to the traditional "hard constraint" version.

## RESEARCH STATUS

### 1. INTRODUCTION

In our problem an optical sensor has a sunshade that limits the field of view. The limited scene is subsequently Fourier transformed and a number of samples is taken in the Fourier domain. These frequency domain measurements are naturally subject to measurement noise. The idea now is to recover as much as possible of the original scene.

Philosophically at least this is a reasonable idea due to the analogy with the problem of extrapolating bandlimited signals. Several iterative projection algorithms have been proposed [2], and proofs of convergence have been given for the continuous problem [3,4]. The successful extrapolation by a factor of 20, in [5], led the author to extend and apply that computationally efficient one-shot approach to the 2-D problem outlined above [1]. The results of that approach were not encouraging due to an extreme sensitivity to noise, as well as the difficulty of implementing all a priori information. An iterative projection algorithm [6] facilitates the incorporation of a priori information at the expense of an increased computational burden. We develop a modification of some of the constraints, to accommodate the effects of measurement noise.

## 2. PROBLEM FORMULATION

As extensions to 2-D are rather elementary, 1-D formulations are used for transparency of the derivations.

Let  $f(x)$  denote the continuous space-limited scene, and  $F_C(\omega)$  its Fourier transform. The sequence  $\{F_n\}$  is obtained from  $F_C(\omega)$  by sampling with interval  $\Delta$ . A periodic function,  $f_s(t)$ , period  $2\pi$ , can be associated with  $\{F_n\}$  as follows

$$f_s(t) = \sum_{n=-\infty}^{\infty} F_n e^{jtn} \quad (1)$$

For the Fourier coefficients we have, therefore,

$$F_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} f_s(t) e^{-jtn} dt \quad (2)$$

As  $\{F_n\}$  was obtained by sampling, the following relationship is valid

$$\begin{aligned} F_n &= F_C(n\Omega) \\ &= \int_{-\infty}^{\infty} f(x) e^{-jn\Omega x} dx \\ &= \sum_{r=-\infty}^{\infty} \int_{(2r-1)\pi/\Omega}^{(2r+1)\pi/\Omega} f(x) e^{-jn\Omega x} dx \end{aligned}$$

Substituting  $\bar{x} + \frac{2\pi r}{\Omega}$  for  $x$  yields

$$F_n = \sum_{r=-\infty}^{\infty} \int_{-\pi/\Omega}^{\pi/\Omega} f(\bar{x} + \frac{2\pi r}{\Omega}) e^{-jn\Omega \bar{x}} e^{-jn\Omega 2\pi r/\Omega} d\bar{x}$$

Recognizing the last exponential to have a power which is an integer multiple of  $2\pi$ , and substituting  $t$  for  $\bar{x}\Omega$ , results in:

$$F_n = \int_{-\pi}^{\pi} \sum_{r=-\infty}^{\infty} f((t+2\pi r)/\Omega) e^{-jnt} dt/\Omega \quad (3)$$

Comparing (2) and (3) yields the relationship between the scene-limited function  $f(\cdot)$  and the periodic function  $f_s(\cdot)$ .

$$f_s(t) = \sum_{r=-\infty}^{\infty} f((t+2\pi r)/\Omega) \quad (4)$$

The corresponding 2-D result follows.

$$f_s(t_1, t_2) = \sum_{r_1=-\infty}^{\infty} \sum_{r_2=-\infty}^{\infty} f((t_1+2\pi r_1)/\Omega_1, (t_2+2\pi r_2)/\Omega_2)$$

Assume that  $f(x)$  is scene-limited to  $[x_1, x_2]$ , an interval of length  $x_2 - x_1$ . From (4) the original function  $f(x)$  can be recovered from the periodic function  $f_s(t)$  associated with the frequency domain samples, if those samples are spaced closely enough, that is

$$\Omega \geq \frac{2\pi}{x_2 - x_1} \quad (5)$$

The important result is, that the space-limited scene can be recovered from all frequency domain samples, under the constraint of (5). In the sequel (5) will be assumed valid by some margin, so called oversampling, so that  $f_s(t)$  is truly scene-limited to a subset of  $[-\pi, \pi]$ .

It is a real world problem that all frequency domain samples are necessary. For the space based infrared sensor this requirement translates into an infinitely large detector array, which is clearly out of the question. In practice, therefore, we hope to get access to the dominant portion

of  $\{F_n\}$ , by means of an extrapolation process, so that

$$f_s(\tau) = \sum_{n=-\infty}^{\infty} F_n \tau^{-n} \quad (6)$$

$$\approx \sum_N F_n \tau^{-n} \quad (7)$$

A uniform sampling of  $f_s(\tau)$  on the unit circle in the complex  $\tau$ -plane, yields an  $N$ -sequence,  $\{f_n\}$  say, for which

$$\begin{aligned} f_k &= f_s(\tau)/\tau = W_N^{-k} \\ &= \sum_{n=-\infty}^{\infty} F_n W_N^{kn} \end{aligned} \quad (8)$$

where

$$W_N \triangleq e^{-j2\pi/N} \quad (9)$$

Now a 1-to-1 relation exists between the  $N$ -sequence  $\{f_n\}$  and the  $N$ -sequence  $\{F_N\}$ .

$$F_n = \frac{1}{N} \sum_{k=0}^{N-1} f_k W_N^{-kn} \quad (10)$$

A substitution of (8) into (10) yields

$$\begin{aligned} F_n &= \frac{1}{N} \sum_{k=0}^{N-1} \left[ \sum_{m=-\infty}^{\infty} F_m W_N^{km} W_N^{-kn} \right] \\ &= \sum_{m=-\infty}^{\infty} F_m \left[ \frac{1}{N} \sum_{k=0}^{N-1} W_N^{-k(n-m)} \right] \\ &= \sum_{m=-\infty}^{\infty} F_m \delta_{m-n-rN} \quad \forall \text{ integer } r \\ &= \sum_{r=-\infty}^{\infty} F_{n+rN} \end{aligned} \quad (11)$$

Consequently (7) will be an increasingly better approximation as  $N$  becomes large enough for the  $r=0$  term in (11) to dominate. The accuracy of the detail in the reconstructed scene depends on  $N$  and the degree to which (7) is satisfied.

A graphical representation of the problem is shown in Figure 1.

### 3. SOLUTION APPROACH

#### A. Alternating Projection Algorithm

For the derivation and analysis that follows (similar to [6]) we assume the discrete frequency sequence  $\{F_n\}_1^N$  and the associated periodic function  $f_s(t)$  over  $[-\pi, \pi]$  to constitute our problem. Only a small number of the frequency domain samples is available due to a frequency domain distortion or truncation operator  $T$ .

$$Y = TF \quad (11)$$

The aim now, is to use all available a priori information to extrapolate, or rather estimate, the frequency domain samples. If such estimation is successful, a better estimate of the original is achieved than would be possible from the measurements without extrapolating estimation.

In addition to (11) a solution  $F$  must satisfy a number of constraints in the scene domain, indicated by a compound operator  $C_S$ . Also a constraint in the frequency domain exists, and is denoted by  $C_F$ . Consequently, we have

$$F = FC_S F^{-1} C_F F \quad (13)$$

or equivalently

$$F = C_F F C_S F^{-1} F \quad (14)$$

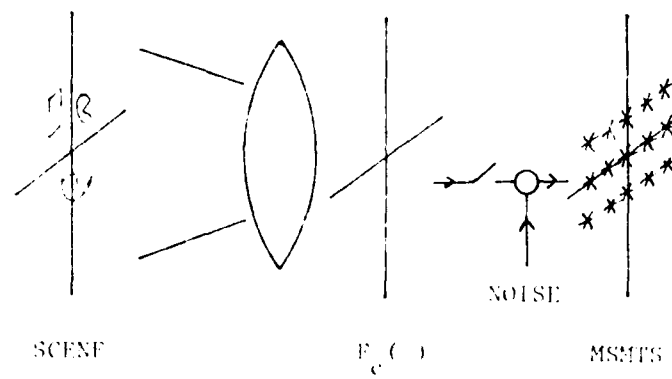


Figure 1. Problem Representation.



For notational convenience, let's define a constraint operator  $C$  encompassing (13) and (14), so that

$$F = CF \quad (15)$$

From (12) and (15) the following equality is valid, for any  $\mu$

$$\begin{aligned} F &= CF + \mu (Y - TCF) \\ &= OF \end{aligned} \quad (16)$$

$$= \mu Y + (I - \mu T)CF \quad (17)$$

The problem is to find a solution  $F$  to the above, or in other words, to find a so-called fixed point of the operator  $C$ .

A common mode of attack is the successive approximation approach, implemented by

$$F_{k+1} = \mu Y + (I - \mu T)CF_k \quad (18)$$

According to the contraction mapping theorem, (18) yields a unique solution if  $C$  is a contraction operator, i.e., if  $0 \leq \alpha \leq 1$  exists, such that

$$\|CF_1 - CF_2\| \leq \alpha \|F_1 - F_2\| \quad \forall F_1, F_2 \quad (19)$$

If  $\alpha$  can take on the unit value, (19) defines a nonexpansive operator. A nonexpansive operator therefore decreases the distance between signals. The equivalent requirement for convergence of (18), is that  $(I - \mu T)C$  is a contraction opera-

tor. For the latter it is sufficient that  $I - \mu I$  and  $C$  are both nonexpansive, and one of the two is a contraction operator. As the second term on the righthand side of (18) constitutes a correction term over the measurement domain, one often encounters the choice  $\mu=1$ , expressing no need for correcting  $\mu Y$  over that domain. We'll see shortly that in the present approach such a choice is not applicable.

#### B. Soft Constraints

One of the most crucial issues in having a chance at establishing some kind of convergence for the iterative process in (18), is the compatibility of the set of constraints that applies to the solution. As physical measurements are invariably subject to noise effects, it seems ill-advised to impose these noisy measurements as a hard constraint, i.e. as if these constituted absolute knowledge. As a matter of fact, such a hard constraint may lead to an incompatible constraint set, as demonstrated by the following example from the spectral estimation arena.

Suppose a small number of low-lag values of a covariance sequence is available, and extrapolation of these is desired in order to increase spectral resolution. A theoretical constraint on the spectral density function would be nonnegativity. Now suppose that due to measurement noise the zero lag covariance value is not the largest in magnitude. This now violates one of the necessary properties for covariance sequences, and results in an incompatibility of the given covariance sequence segment with any spectral density function.

In order to accommodate errors in the available measurements, we introduce the following soft constraint as frequency domain measurement constraint.

$$C_F F = \begin{cases} F & \text{if } ||F-Y||_M = \delta \leq \varepsilon \\ Y + \frac{\varepsilon}{\delta} (F-Y) & \text{over } M \\ F & \text{over } M^C \end{cases} \quad \text{if } ||F-Y||_M \neq 0 \quad (20)$$

Herein,  $\varepsilon^2$  indicates the tolerable, and  $\delta^2$  indicates the actual, mean square difference between reconstruction and measurements. These evaluations are made over the measurement domain  $M$  only. If corrective action takes place, the following relationship holds

$$\begin{aligned} ||C_F F - Y||_M &= ||Y + \frac{\varepsilon}{\delta} (F-Y) - Y||_M \\ &= \frac{\varepsilon}{\delta} ||F-Y||_M = \varepsilon \end{aligned}$$

Therefore, we find,

$$\begin{aligned} ||C_F F - Y||^2 &= ||F-Y||_{M^C}^2 + ||C_F F - Y||_M^2 \\ &= ||F-Y||_{M^C}^2 + \frac{\varepsilon^2}{\delta^2} ||F-Y||_M^2 \\ &\leq ||F-Y||^2 \end{aligned} \quad (21)$$

which says that the iterate either remains unchanged, or moves closer to the measurements. Our soft measurement constraint leaves well enough alone, i.e. if the iterative process comes close enough to  $Y$ , with respect to the expect-

ed noise level, then the iterate is considered acceptable. An alternative soft constraint would be one that places a tolerated maximum on deviations from the measurements, for each measurement sample individually. Note that in any case a weighted norm is easily accommodated, so that known amplitude-and/or frequency-dependent noise information may readily be incorporated.

Several constraints apply in the scene domain. As the solution to our problem is a picture, compatible with our information, a nonnegativity operator applies, which is not softened. The scene is also of limited extent and it is argued that softening may be beneficial here. A border region B is proposed (similarities with [7]), to serve as a transition region between the known extent K, and the zeroed region Z. The soft scene limitation operator S then becomes

$$SF^{-1}F = \begin{cases} f^{-1}F & \text{over } K \\ g_B f^{-1}F & \text{over } B \\ 0 & \text{over } Z \end{cases} \quad (22)$$

where  $0 \leq g_B < 1$  is a function defined over the border region, which provides a smooth transition to zero.

In the border region itself one could elect not to make any substitutions, and then monitor the energy in the border region. This information can potentially be used to accelerate the convergence of the algorithm. We hope to investigate this issue in the subsequent stage of research.

### C. Convergence Issues

In order to assess the prospects for convergence of (13) the contraction properties of several operators must be evaluated [6,8]. For the scene limitation operator  $S$ , via Parseval's relationship

$$\begin{aligned} ||SF^{-1}F_i - SF^{-1}F_j||^2 &= \frac{1}{2\pi} \int_K |f_i - f_j|^2 dt + \frac{1}{2\pi} \int_B |g_B|^2 |f_i - f_j|^2 dt \\ &= ||F^{-1}F_i - F^{-1}F_j||^2 \cdot r^2 \end{aligned} \quad (23)$$

where

$$r^2 = 1 - \frac{\frac{1}{2\pi} \int_Z |f_i - f_j|^2 dt + \frac{1}{2\pi} \int_B (1 - |g_B|^2) |f_i - f_j|^2 dt}{||F^{-1}F_i - F^{-1}F_j||^2} \quad (24)$$

The expression for  $r^2$  is nonnegative and smaller or equal one, where equality occurs if  $f_i$  and  $f_j$  are equal over the union of  $B$  and  $Z$ . The operator  $S$  is therefore nonexpansive.

The nonnegativity operator can similarly be shown to be nonexpansive.

$$\begin{aligned} ||PF^{-1}F_i - PF^{-1}F_j||^2 &= \frac{1}{2\pi} \int_{-\pi}^{\pi} |Pf_i - Pf_j|^2 dt \\ &\leq \frac{1}{2\pi} \int_{-\pi}^{\pi} |f_i - f_j|^2 dt \\ &\leq ||F^{-1}F_i - F^{-1}F_j||^2 \end{aligned} \quad (25)$$

Equality holds if the functions  $\text{sign}(f_i)$  and  $\text{sign}(f_j)$  are equal almost everywhere  $[-\pi, \pi]$ . The composite scene domain operator

$$C_S = PS$$

$$= SP \quad (26)$$

is, consequently, a nonexpansive operator.

The soft frequency domain measurement constraint yields,

$$\|C_F F_i - C_F F_j\|^2 = \|C_F F_i - C_F F_j\|_M^2 + \|C_F F_i - C_F F_j\|_{M^C}^2 \quad (27)$$

$$\leq \|F_i - F_j\|_M^2 + \|F_i - F_j\|_{M^C}^2 \quad (28)$$

$$\leq \|F_i - F_j\|^2 \quad (29)$$

The inequality arises because the constraint leaves the components of  $F_i$  and  $F_j$  in  $M$  alone, or moves one or both components toward  $Y$ , and consequently closer together. The components in  $M^C$  are always unchanged. We find the measurement operator  $C_F$ , therefore, to be a nonexpansive operator also.

Rests us to evaluate the operator  $(I-uI)$ .

$$\|F_i - uIF_i - F_j + uIF_j\|^2 = \|F_i - uF_i - F_j + uF_j\|_M^2 + \|F_i - F_j\|_{M^C}^2 \quad (30)$$

$$= (1-u)^2 \|F_i - F_j\|_M^2 + \|F_i - F_j\|_{M^C}^2 \quad (31)$$

$$\leq \|F_i - F_j\| \quad \text{for } 0 \leq u \leq 2 \quad (32)$$

Equality holds if  $F_i$  equals  $F_j$  over  $M$ .

The conditions for equality in (24), (25), (29), and (32) could all be met at once, and consequently we find  $(I-uI)C$

to be a non-expansive operator. As a result (18) may have many fixed points. It should be recognized that in view of the noisy measurements, a unique solution cannot be expected. Any solution, however, must satisfy all the constraints. The size of the solution set can possibly be reduced by decreasing  $\epsilon$  in (20), but this moves the problem towards incompatibility of the available information.

#### D. Algorithm Simplification

Using the soft frequency domain constraint in (20), and the composite scene domain constraint in (26), in the algorithm formulation of (18), yields the following algorithm.

$$F_0 = \mu Y$$

$$F_{k+1} = \mu Y + (1-\mu) C_F^T C_S F^{-1} F_k \quad (33)$$

This algorithm is represented in Figure 2.

Let's define

$$\tilde{F}_k = C_F^T C_S F^{-1} F_k \quad (34)$$

so that the right hand side of (33) can be written and simplified as follows.

$$\tilde{F}_k - \mu \tilde{F}_k + \mu Y = \tilde{F}_k \quad \text{over } M^C \quad (35)$$

$$= (1-\mu) \tilde{F}_k + \mu Y \quad \text{over } M \quad (36)$$

The latter equality can be rewritten as

$$\tilde{F}_k - \mu \tilde{F}_k + \mu Y = Y + (1-\mu)(\tilde{F}_k - Y) \quad (37)$$

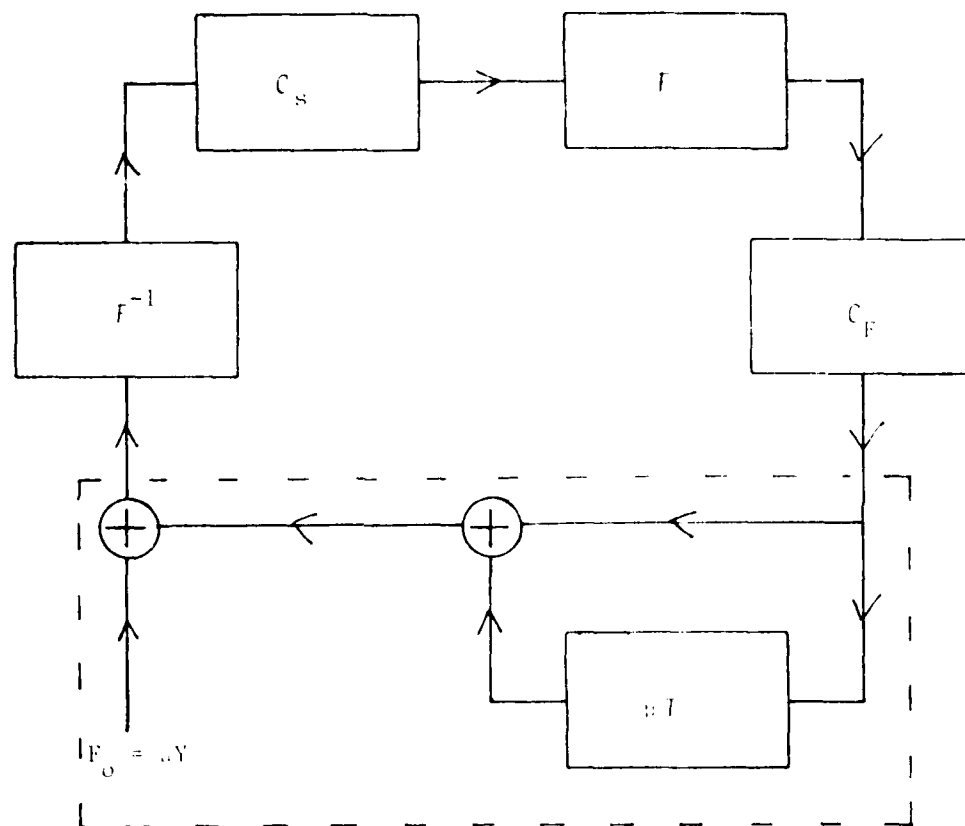


Figure 2. Algorithm Simplification.



If we now substitute  $\frac{\epsilon}{\delta}$  for  $(1-\mu)$  then (35) and (37) can be seen to represent the soft frequency domain constraint of (20). As  $\frac{\epsilon}{\delta}$  is smaller or equal one the corresponding choice for  $\mu$ :

$$\mu = 1 - \frac{\epsilon}{\delta} \quad (38)$$

is seen to satisfy the convergence conditions of (32). As a result of making this particular choice for  $\mu$ , we would have in Figure 2, two identical consecutive frequency domain constraint operators. The algorithm therefore simplifies because the dot-enclosed algorithm part now has the effect of an identity operator, due to the equality

$$C_F \cdot C_F = I \cdot C_F \quad (39)$$

The resulting algorithm simply consists of transformations from frequency domain to scene domain and vice versa, with an application of the respective constraint operators.

Note that if our tolerable noise level equals zero, i.e.  $\epsilon = 0$  in (20), then (20) as well as (38) cause the algorithm to degenerate to the traditional hard-constraint algorithm, in which the iterate is replaced with the measurements over the measurement domain.

#### 4. SIMULATIONS

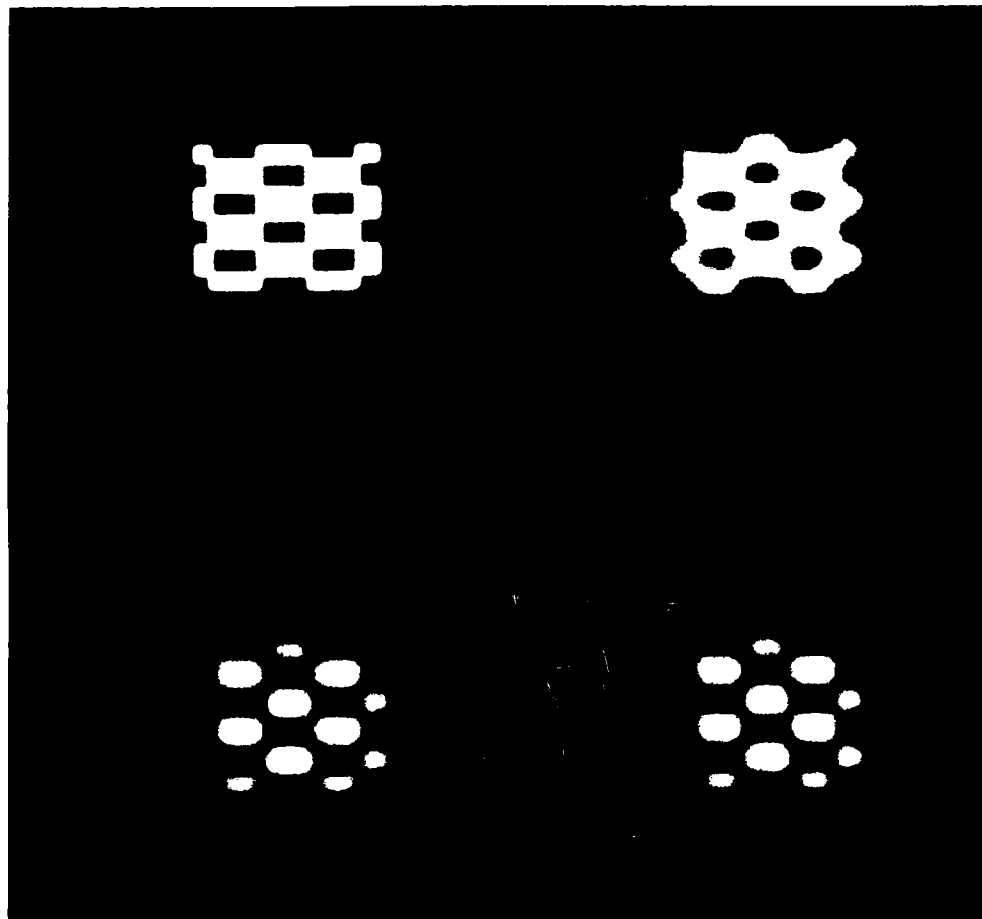
For each of the blocks in Figure 2 commands were written in RATFOR, for subsequent running on the VAX 11/780 in VPI&SU's SDA lab. A command is like a self-contained subroutine called from the keyboard, and usually operating on existing files. Each iteration of the algorithm therefore takes four commands. One more command was written to take care of all initializations. Codes for these commands can be found in the Appendix.

Our scene domain consists of a  $64 \times 64$  pixel array. After generating a checkerboard pattern the scene domain was constrained to the center  $25 \times 20$  (hor x vert) pixels, which therefore constitute the region of known extent K. The resulting scene limited picture was then transformed and the measurement domain M was chosen to be the  $15 \times 19$  (hor x vert) area centered at zero frequency. The mean square value of the frequency domain signal turned out to be  $27 \times 10^6$ . Noise can be added to these frequency domain measurement values. We used a noise variance of  $40 \times 10^3$ . The scene domain constraint implemented in these simulations left a scene of  $27 \times 22$  (hor x vert) pixels, so that the border region B is 1 pixel wide on each side.

At present the set of commands has to be entered sequentially in order to effect a number of iterations of the algorithm. This time consuming process can be alleviated in the future by rewriting the commands into a

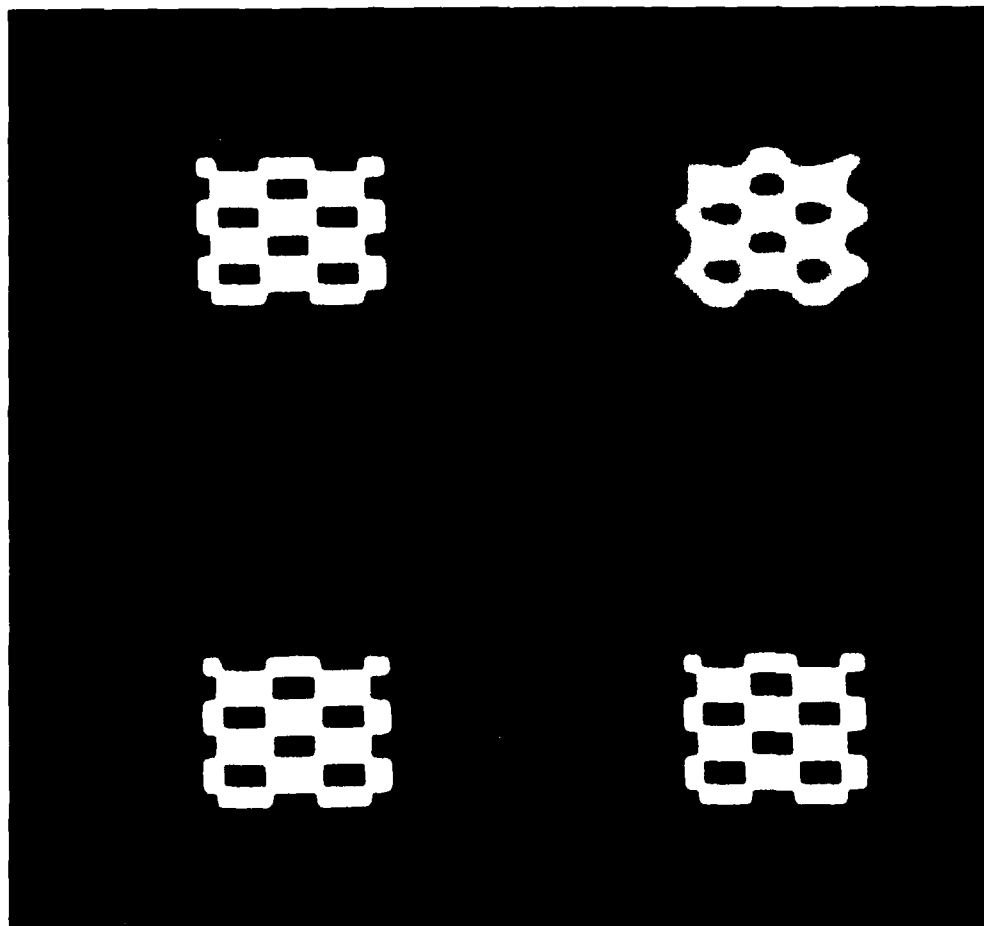
single command, which will then facilitate a more comprehensive study of the convergence behavior for large numbers of iterations of the present algorithm. Consequently our present numerical results are limited to 7 iterations of the algorithm. This limited experiment, however, demonstrates what is expected to be the characteristic behavior of the algorithm.

Starting from the initial condition without noise and going through 7 iterations will not lead to any dramatic results. We illustrate, nevertheless, the operational condition of the algorithm in Figure 3, where a substantial improvement in resolution is obtained after only seven iterations, when compared to a simple inverse transform of the measurements. To illustrate the characteristic behavior of the soft-constraint algorithm on the basis of only 7 iterations, the next experiment, with noisy measurements (SNR = 28 dB) starts from the other end. Let the true solution be the initial iterate, which should therefore constitute a feasible solution for the algorithm. The iterate  $F_k$  in the traditional algorithm with the hard frequency domain constraint then moves away from the true solution  $F$ , thereby indicating that this algorithm cannot possibly converge to the true solution. This effect was actually observable on the TV monitor, although it does not show in the photograph of Figure 4. We therefore derived numerical indicators of the behavior of the reconstructions,



Upper Left: Original Scene  
Upper Right: Inverse of Measurements without Noise  
Lower Left: Iterate 7 before Scene Domain Constraint Operator  
Lower Right: Iterate 7 after Scene Domain Constraint Operator

Figure 3. Algorithm Operation for Noiseless Measurements,  
with Inverse of Measurements as Initial Iterate.



Upper Left: Original Scene  
Upper Right: Inverse of Measurements with Noise  
Lower Left: Iterate 7 before Scene Domain Constraint Operator  
Lower Right: Iterate 7 after Scene Domain Constraint Operator

Figure 4. Algorithm Operation for Noisy Measurements,  
with Original Scene as Initial Iterate.

and these RMS values are represented in Figure 5. The iterate in the soft constraint algorithm with the noise variance underestimated at 10,000 (the noise variance equals 40,000) is seen to also move away from the true solution, but much slower. Finally, it is clear that if a good estimate of the noise variance is available, then the iterate will not move away from the true solution, because they are close enough already to satisfy the soft constraint. The specific effects of estimated and actual noise variance will be investigated in a subsequent research effort.

## REFERENCES

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7. T. F. Quatieri, D. E. Dudgeon, "Implementation of 2-D Digital Filters by Iterative Methods," ASSP-30, No. 3, June 1982.
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```

#--FFTDRV                                DRIVER
#
#IDENTIFICATION
#      TITLE                                FFTDRV
#      AUTHOR                               AA(LOUIS) BEEH
#      VERSION                              A.01
#      DATE                                 15 SEPT 1982
#      LANGUAGE                             RATFOR
#      SYSTEM                               VAX-11
#
#PURPOSE
#      DRIVER FOR COMMAND THAT PERFORMS THE RADIX-2 FAST
FOURIER #      TRANSFORM OF A REAL IMAGE
#
#ENTRY POINT
#      FFTDRV(WORK,ERRET)
#
#ARGUMENT LISTING
#      WORK                                INT                WORK ARRAY
#      ERRET                               INT                ALTERNATE ERROR RETURN
#
#INCLUDE FILES/COMMONS
#      MACAL                               INCLUDE            GIPSY TOKEN DEFINITIONS FOR A1
CHARACTER #      GIPCOM                     INCLUDE
#      ERRET                               INCLUDE            INCLUDE FILE FOR COMMON ERROR
#
#ROUTINES CALLED
#      PPUSH                               PUSHES PROGRAM NAME INTO ERROR STACK
(GIPSY) #      PPOP                               POPS PROGRAM NAME FROM ERROR STACK
(GIPSY) #      RDKINL                             OPEN A SIF AND FILL IN THE IDENT
BLOCK #      CLOSE                               CLOSSES A FILE(GIPSY)
#      FFTALG                               COMPUTES THE (INVERSE) FAST FOURIER
TRANSFORM (USER) #

```

```

*****
#
INCLUDE MACAL
SUBROUTINE FFTDRV(WORK,*)
IMPLICIT INTEGER (A-Z)
INCLUDE GIPCOM
INCLUDE ERROR
INTEGER IDENT(.IDLENGTH)
DIMENSION WORK(.ARB)
#

```



[illegible]

1

```
#
IEV=-2012
GOTO 9999
#
9010 CONTINUE
IEV=OSGIEV(IEV)
GOTO 9999
#
#
9020 CONTINUE
IEV=-5004
GOTO 9999
#
9999 CONTINUE
CALL CLOSE(FDI1)
CALL CLOSE(FDO1)
#
RETURN 1
END
```

ILLEGAL DATA MODE

ILLEGAL ARRAY SIZE

```

#--FFTALG                                I/O &ALGORITHM
#
#IDENTIFICATION
#      TITLE                FFTALG
#      AUTHOR               AA(LOUIS) BEEK
#      VERSION              A.01
#      DATE                 15 SEPT 1982
#      LANGUAGE             RATFOR
#      SYSTEM               VAX-11
#
#PURPOSE
#      TO COMPUTE THE FAST FOURIER TRANSFORM (RADIX-2)
#      OF A SIF IMAGE (1 BND IF REAL, 2 BNDS IF COMPLEX)
#
#ENTRY POINT
#
FFTALG(FDI,FDO,NBND,ILIN1,ILIN2,XAR,FTAR,NLIN,NPPL,NMAX,IEV,*) #
#ARGUMENT LISTING
#      FDI                  INPUT FILE DESCRIPTOR
#      FDO                  OUTPUT FILE DESCRIPTOR
#      NBND                 NO OF BANDS IN IMAGE TO BE PROCESSED
#      ILIN1                LINE BUFFER FOR REAL PART OF INPUT IMAGE
#      ILIN2                LINE BUFFER FOR IMAGINARY PART OF IMAGE
#      XAR                  WORKARRAY
#      FTAR                 WORKARRAY
#      NLIN                 NO OF ROWS IN THE IMAGE
#      NPPL                 NO OF COLUMNS IN THE IMAGE
#      NMAX                 MAX OF NPPL AND NLIN
#      INV                  INVERSE TRANSFORM IF TRUE
#      IEV                  INTEGER EVENT VARIABLE
#      ERRET                ALTERNATE RETURN
#
#INCLUDE FILES/COMMONS
#      MACA1                INCLUDE (GIPSY)
#
#ROUTINES CALLED
#      PPUSH
#      PPOP
#      CLOSE
#      CPYIDK
#      COPYDS
#      RREAD
#      RWRITE
#      RX2FFT
#      DSCNAM
#
*****
#
  INCLUDE MACA1
  SUBROUTINE
    FFTALG(FDI1,FDO1,NBND,ILIN1,ILIN2,XAR,FTAR,NLIN,NPPL,NMAX,IEV,*
    IMPLICIT INTEGER (A-Z) CHARACTER FDI1(.FDLENGTH),
    FDO1(.FDLENGTH) INTEGER IDENT(.IDLENGTH), JDENT(.IDLENGTH)
    REAL ILIN1(NPPL), ILIN2(NPPL), NORM
    COMPLEX XAR(NMAX), FTAR(NLIN,NPPL)
    COMPLEX Z
    LOGICAL INV
  #
  INCLUDE TTCOM
  #

```

```

#
#
CALL PPUSH('FFTALG')
#
NMAX=MAX(NLIN,NPPL)
FORWARD=1
INVERSE=-1
DFLT=1
CALL RNGETI('FORWARD(1) OR INVERSE(-1)
TRANSFORM?.',INVERSE,FORWARD,DFLT, VAR,IEV,%9000)
INV=.TRUE.
IF(VAR==1) INV=.FALSE.
#
OPEN INPUTFILE
CALL CPYIDR(FDI1,IDENT,.OPNTMP,IEV,%9000)
#
#
WRITE DESCRIPTOR RECORD TO TEMP
FILE CALL DSCNAM('FFTALG',IEV,%9000)
#
#
SET UP OUTPUT FILE PARAMETERS
JDENT(.IDNPPL)=IDENT(.IDNPPL)
JDENT(.IDNLINS)=IDENT(.IDNLINS)
JDENT(.IDNCOLS)=IDENT(.IDNCOLS)
JDENT(.IDNROWS)=IDENT(.IDNROWS)
JDENT(.IDNBND)=2
JDENT(.IDMODE)=2
#
#AN INVERSE TRANSFORM MUST YIELD A REAL IMAGE FOR OUR
APPLICATION: #NO. OF OUTPUT BANDS EQUALS 1
NOB=2
IF(INV) NOB=1
JDENT(.IDNBND)=NOB
#
#
OPEN OUTPUT FILE
CALL COPYDS(FDO1,JDENT,IEV,%9000)
#
DO BLKN=1,NLIN
$(
CALL RREAD(FDI1,ILIN1,1,BLKN,IDENT,.WAIT,IEV,%9000)
IF(NBND==2)
$(
CALL RREAD(FDI1,ILIN2,2,BLKN,IDENT,.WAIT,IEV,%9000)
$)
ELSE
$(
DO I=1,NPPL
$(
ILIN2(I)=0.
$)
$)
IF(INV)
$(
DO J=1,NPPL
$(
XAR(J)=CMPLX(ILIN1(J),ILIN2(J))
XAR(J)=CONJG(XAR(J))
$)
$)
ELSE

```

```

      $(
      DO J=1,NPPL
        $(
          XAR(J)=CMPLX(ILIN1(J),ILIN2(J))
        $(
      $(
      CALL RX2FFT(XAR,NPPL)
      DO J=1,NPPL
        $(
          FTAR(BLKN,J)=XAR(J)
        $(
      $(
      NORM=FLOAT(NLIN*NPPL)
      DO J=1,NPPL
        $(
          DO I=1,NLIN
            $(
              XAR(I)=(FTAR(I,J))
            $(
          CALL RX2FFT(XAR,NLIN)
          IF(INV)
            $(
              DO I=1,NLIN
                $(
                  FTAR(I,J)=CONJG(XAR(I))/NORM
                $(
              $(
            ELSE
              $(
              DO I=1,NLIN
                $(
                  FTAR(I,J)=XAR(I)
                $(
              $(
            $(
          #
          DO BLKN=1,NLIN
            $(
              DO I=1,NPPL
                $(
                  Z=FTAR(BLKN,I)
                  ILIN1(I)=REAL(Z)
                  ILIN2(I)=AIMAG(Z)
                $(
              CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
              IF(NOBS==1) GOTO 1000
              CALL RWRITE(FD01,ILIN2,2,BLKN,JDENT,.WAIT,IEV,%9000)
              1000 CONTINUE
            $(
          CALL PPOP
          CALL CLOSE(FDI1)
          CALL CLOSE(FD01)
          RETURN
          #
          #
          #
          9000 CONTINUE
          CALL CLOSE(FDI1)

```

ERROR CONDITIONS

CALL CLOSE(FD01)

#

9999 CONTINUE

9090 RETURN 1

END

```

#--RX2FFT      THIS IS A SUBROUTINE CALLED IN FFTALG.RAT
SUBROUTINE RX2FFT(X,N)
COMPLEX X(N), U, W, T
NV2=N/2
NM1=N-1
M=1
WHILE(NV2>1)
$(
M=M+1
NV2=NV2/2
$)
NV2=N/2
J=1
DO I=1,NM1
$(
IF(I>=J) GOTO 5
T=X(J)
X(J)=X(I)
X(I)=T
5 K=NV2
6 IF(K>=J) GOTO 7
J=J-K
K=K/2
GOTO 6
7 J=J+K
$)
PI=4.*ATAN(1.)
DO L=1,M
$(
LE=2**L
LE1=LE/2
U=CMPLX(1.,0.)
FLE1=FLOAT(LE1)
CARG=COS(PI/FLE1)
SARG=-1.*SIN(PI/FLE1)
W=CMPLX(CARG,SARG)
DO J=1,LE1
$(
DO I=J,N,LE
$(
IP=I+LE1
T=X(IP)*U
X(IP)=X(I)-T
X(I)=X(I)+T
$)
U=U*W
$)
$)
RETURN
END

```

---

This subroutine is an adaptation of the Cooley, Lewis, and Welch algorithm for decimation-in-time, radix-2, in-place FFT.

```

#--SCNCTD                                DRIVER
#
#IDENTIFICATION BEEX, RATFOR, VAX-11
#
#PURPOSE
#           DRIVER FOR COMMAND THAT ZEROES A SCENE
#           OUTSIDE OF A SPECIFIED WINDOW
#
#ENTRY POINT SCNCTD(WORK,ERRET)
#
#INCLUDE FILES/COMMONS
#           MACAL
#           GIPCOM
#           ERRET
#
#ROUTINES CALLED
#           PPUSH
#           PPOP
#           RDKINL
#           CLOSE
#           SCNCTA      ALGORITHM SECTION OF SCNCTT COMMAND
#
#*****
#
#INCLUDE MACAL
#SUBROUTINE SCNCTD(WORK,*)
#IMPLICIT INTEGER (A-Z)
#INCLUDE GIPCOM
#INCLUDE ERROR
#INTEGER IDENT(.IDLENGTH)
#DIMENSION WORK(.ARB)
#
#EQUIVALENCE (NPPL,IDENT(.IDNPPL))
#EQUIVALENCE (NLIN,IDENT(.IDNLINS))
#EQUIVALENCE (NCOL,IDENT(.IDNCOLS))
#EQUIVALENCE (NROW,IDENT(.IDNROWS))
#EQUIVALENCE (NBND,IDENT(.IDNBDS))
#EQUIVALENCE (MODE,IDENT(.IDMODE))
#
#CALL PPUSH('SCNCTD')
#
#           OPEN INPUTFILE
#
#CALL RDKINL(FDI1,IDENT,.OLD,IEV,%9999)
#CALL CLOSE(FDI1)
#
#           CHECK INPUTFILE
#
#IF (MODE^=.REALMODE) GOTO 9000
#
#NP=NPPL*NLIN
#N=1
#NP2=NP/2
#WHILE(NP2>1)
#  $(
#    N=N+1
#    NP2=NP2/2
#  $)
#IF (NP^=2**N) GOTO 9020
#IF (NBND^=1) GOTO 9020
#
#NXT=1

```



```

ILIN1=GETWP(NXT,.REALMODE,NPPL)
IF(.OK^=OSALOC(NXT)) GOTO 9010
#
CALL CONTIN(%9999)
#
#           CALL ALGORITHM SECTION
#
CALL SCNCTA(FDI1,FD01,WORK(ILIN1),NPPL,NLIN,IEV,%9999)
#
CALL PPOP
#
RETURN
#           ERROR CONDITIONS
9000 CONTINUE
#           ILLEGAL DATA MODE
IEV=-2012
GOTO 9999
#
9010 CONTINUE
IEV=OSGIEV(IEV)
GOTO 9999
#           ILLEGAL ARRAY SIZE
9020 CONTINUE
IEV=-5004
GOTO 9999
#
9999 CONTINUE
CALL CLOSE(FDI1)
CALL CLOSE(FD01)
#
RETURN1
END

```

```

#--SCNCTA                                ALGORITHM
#
#IDENTIFICATION BEEH, RATFOR, VAX-11
#
#PURPOSE
#           ALGORITHM SECTION FOR COMMAND THAT ZEROES
#           A SCENE OUTSIDE OF A SPECIFIED WINDOW
#
#ENTRY POINT
#           SCNCTA(FDI,FDO,ILIN1,NPPL,NLIN,IEV,*)
#
#ARGUMENT LISTING
#           FDI    INPUTFILE DESCRIPTOR
#           FDO    OUTPUTFILE DESCRIPTOR
#           ILIN1  LINE BUFFER
#           NPPL   NO POINT PER LINE
#           NLIN   NO LINES
#           IEV
#
#*****
#
#INCLUDE MACAL
#SUBROUTINE SCNCTA(FDI1,FDO1,ILIN1,NPPL,NLIN,IEV,*)
#IMPLICIT INTEGER (A-Z)
#CHARACTER FD1(.FDLENGTH), FDO1(.FDLENGTH)
#INTEGER IDENT(.IDLENGTH), JDENT(.IDLENGTH)
#REAL ILIN1(NPPL)
#
#INCLUDE TCOM
#
#CALL PPUSH('SCNCTA')
#
#HLOW=2
#HHI=NPPL-1
#DFLT=NPPL/10
#CALL RNGETI('HORIZONTAL WINDOW
#SIZE?.',HLOW,HHI,DFLT,WXLEN,IEV,%9000)  VLO=2
#VHI=NLIN-1
#DFLT=NLIN/10
#CALL RNGETI('VERTICAL WINDOW
#SIZE?.',VLO,VHI,DFLT,WYLEN,IEV,%9000)  IF(WXLEN^=(WXLEN/2)*2)
#WXLEN=WXLEN+1  IF(WYLEN^=(WYLEN/2)*2) WYLEN=WYLEN+1
#
#           OPEN INPUTFILE
#
#CALL CPYIDR(FDI1,IDENT,.OPNTMP,IEV,%9000)
#CALL DSCNAM('SCNCTA',IEV,%9000)
#
#DO I=6,19
#$(
#IF (I .EQ. 6 .OR. I .EQ. 7 .OR. I .EQ. 13 .OR. I .EQ. 14 .OR.
#   I .EQ. 17 .OR. I .EQ. 19) $(
#           X=IDENT(I)
#           JDENT(I)=X
#           $)
#
#           OPEN OUTPUTFILE
#CALL COPYDS(FDO1,JDENT,IEV,%9000)
#
#IF (WXLEN<=0/WXLEN>=NPPL/WYLEN<=0/WYLEN>=NLIN) GOTO 9000
#
#XWML=(NPPL-WXLEN)/2

```



```

# --FRQCTD                      DRIVER
#
# IDENTIFICATION BEEH, RATFOR, VAX-11
#
# PURPOSE
#     DRIVER SECTION FOR COMMAND THAT IMPOSES A
#     FREQUENCY DOMAIN MEASUREMENT CONSTRAINT
#
# ENTRY POINT FRQCTD(WORK,ERRET)
#
# INCLUDE FILES/COMMONS
#     MACAL
#     GIPCOM
#     ERRET
#
# ROUTINES CALLED
#     PPUSH
#     PPOP
#     RDKINL
#     CLOSE
#     FRQCTA  ALGORITHM SECTION OF FRQCTC COMMAND
# *****
#
# INCLUDE MACAL
# SUBROUTINE FRQCTD(WORK,*)
# IMPLICIT INTEGER(A - Z)
# INCLUDE GIPCOM
# INCLUDE ERROR
# INTEGER IDENT(.IDLENGTH), LDENT(.IDLENGTH)
# DIMENSION WORK(.ARB)
#
# EQUIVALENCE (NPPL,IDENT(.IDNPPL))
# EQUIVALENCE (NLIN,IDENT(.IDNLINS))
# EQUIVALENCE (NCOL,IDENT(.IDNCOLS))
# EQUIVALENCE (NROW,IDENT(.IDNROWS))
# EQUIVALENCE (NBND,IDENT(.IDNBND))
# EQUIVALENCE (MODE,IDENT(.IDMODE))
#
# CALL PPUSH('FRQCTD')
#
#     OPEN INPUTFILES
# CALL RDKINL(FDI1,IDENT,.OLD,IEV,%9999)
# CALL CLOSE(FDI1)
# CALL RDKINL(FDI2,LDENT,.OLD,IEV,%9999)
# CALL CLOSE(FDI2)
#
#     CHECK INPUTFILES
# IF(MODE^=.REALMODE) GOTO 9000
# NEED TO CHECK EQUALITY OF IDENT AND LDENT ARRAYS BUT NOT ALL
#
# NXT=1
# ILIN1=GETWP(NXT,.REALMODE,NPPL)
# ILIN2=GETWP(NXT,.REALMODE,NPPL)
# CORR=GETWP(NXT,.REALMODE,NLIN*NPPL)
# CORI=GETWP(NXT,.REALMODE,NLIN*NPPL)
#
# IF(.OK^=OSALOC(NXT)) GOTO 9010
#
# CALL COMTIN(%9999)
#
# CALL FRQCTA(FDI1,FDI2,FDI1,WORK(ILIN1),WORK(ILIN2),
#             WORK(CORR),WORK(CORI),NLIN,NPPL,IEV,%9999)
#
# CALL PPOP

```

#  
RETURN  
#  
9000 CONTINUE  
#  
IEV=-2012  
GOTO 9999  
#  
9010 CONTINUE  
IEV=OSGIEV(IEV)  
GOTO 9999  
#  
9999 CONTINUE  
CALL CLOSE(FDI1)  
CALL CLOSE(FDI2)  
CALL CLOSE(FDOI)  
#  
RETURN 1  
END

ERROR CONDITIONS  
ILLEGAL DATAMODE/INCOMPATIBLE INPUT ARRAYS

```

#--FRQCTA          I/O & ALGORITHM
#
#IDENTIFICATION BEEX, RATFOR, VAX-11
#
#PURPOSE
#   ALGORITHM AND INPUT/OUTPUT SECTION FOR COMMAND THAT
#   IMPOSES A FREQUENCY DOMAIN MEASUREMENT CONSTRAINT
#
#ENTRY POINT
#   FRQCTA(FDI1,FDI2,FD01,ILIN1,ILIN2,CORR,CORI,NLIN,NPPL,IEV,*)
#
#ARGUMENT LISTING
#   FDI1   MEASUREMENTS
#   FDI2   OLD RECONSTRUCTION
#   FD01   NEW RECONSTRUCTION
#
#INCLUDE FILES/COMMONS
#   MACAL
#
#ROUTINES CALLED
#   PPUSH
#   PPOP
#   CLOSE
#   CPYIDR
#   COPYDS
#   RREAD
#   RWRITE
#   DSCNAM
#
#*****
#
#INCLUDE MACAL
#SUBROUTINE FRQCTA(FDI1,FDI2,FD01,ILIN1,ILIN2,CORR,CORI,
#                  NLIN,NPPL,IEV,*)
#IMPLICIT INTEGER (A - Z)
#CHARACTER FDI1(.FDLENGTH), FDI2(.FDLENGTH), FD01(.FDLENGTH)
#INTEGER IDENT(.IDLENGTH), LDENT(.IDLENGTH), JDENT(.IDLENGTH)
#REAL ILIN1(NPPL), ILIN2(NPPL), SUM, THRESH, SOFCOR, DIF
#REAL CORR(NLIN,NPPL), CORI(NLIN,NPPL)
#REAL TL, TH, TD
#
#INCLUDE TTCOM
#
#CALL PPUSH('FRQCTA')
#
#                               OPEN INPUTFILES
#CALL CPYIDR(FDI1,IDENT,.OPNTMP,IEV,%9000)
#CALL CPYIDR(FDI2,LDENT,.OPNTMP,IEV,%9000)
#                               WRITE DESCRIPTOR RECORD TO TEMP FILE
#CALL DSCNAM('FRQCTA',IEV,%9000)
#CALL DSCNAM('FRQCTA',IEV,%9000)
#                               SET UP OUTPUTFILE PARAMETERS
#DO I=6,19
#$(
#   IF (I .EQ. 6 .OR. I .EQ. 7 .OR. I .EQ. 13 .OR. I .EQ. 14 .OR.
#       I .EQ. 17 .OR. I .EQ. 19) $(
#                               X=IDENT(I)
#                               JDENT(I)=X
#                               $)
#)
#
#                               OPEN OUTPUTFILE
#CALL COPYDS(FD01,JDENT,IEV,%9000)

```

```

#
BLKN=NLIN/2+1
#
#       READ THE INFO PASSED ALONG IN THE MSMT ARRAY
#       LINES 1 THRU YS AND NLIN-YS+2 THRU NLIN FOR
#       COLS 1 THRU XS AND NPPL-XS+2 THRU NPPL
#       CONSTITUTE THE MSMT DOMAIN IN FREQUENCY SPACE
CALL RREAD(FDI1,ILIN1,1,BLKN,IDENT,.WAIT,IEV,%9000)
YS=INT(ILIN1(1))
XS=INT(ILIN1(2))
#
#       DETERMINE MEAN SQUARE DEVIATION OVER MSMT DOMAIN
#       (EVENTUALLY ADD MAXIMUM DEVIATION OPTION ETC)
DO I=1,NLIN
$(
  DO J=1,NPPL
  $(
    CORR(I,J)=0.
    CORI(I,J)=0.
  $)
$)
SUM=0.
DO BLKN=1,YS
$(
  DO BND=1,2
  $(
    CALL RREAD(FDI1,ILIN1,BND,BLKN,IDENT,.WAIT,IEV,%9000)
    CALL RREAD(FDI2,ILIN2,BND,BLKN,IDENT,.WAIT,IEV,%9000)
    DO I=1,XS
    $(
      DIF=ILIN1(I)-ILIN2(I)
      SUM=SUM+DIF**2
      IF(BND==1) CORR(BLKN,I)=DIF
      IF(BND==2) CORI(BLKN,I)=DIF
    $)
    IF(XS>=1)
    $(
      NS=NPPL-XS+2
      DO I=NS,NPPL
      $(
        DIF=ILIN1(I)-ILIN2(I)
        SUM=SUM+DIF**2
        IF(BND==1) CORR(BLKN,I)=DIF
        IF(BND==2) CORI(BLKN,I)=DIF
      $)
    $)
  $)
$)
IF(YS>=1)
$(
  MS=NLIN-YS+2
  DO BLKN=MS,NLIN
  $(
    DO BND=1,2
    $(
      CALL RREAD(FDI1,ILIN1,BND,BLKN,IDENT,.WAIT,IEV,%9000)
      CALL RREAD(FDI2,ILIN2,BND,BLKN,IDENT,.WAIT,IEV,%9000)
      DO I=1,XS
      $(

```

```

DIF=ILIN1(I)-ILIN2(I)
SUM=SUM+DIF**2
IF(BND==1) CORR(BLKN,I)=DIF
IF(BND==2) CORI(BLKN,I)=DIF
$)
IF(XS>=1)
$(
DO I=NS,NPPL
$(
DIF=ILIN1(I)-ILIN2(I)
SUM=SUM+DIF**2
IF(BND==1) CORR(BLKN,I)=DIF
IF(BND==2) CORI(BLKN,I)=DIF
$)
$)
$)
$)
# MEAN SQUARE DIFFERENCE OVER MSMT DOMAIN
SUM=SUM/((2*XS-1)*(2*YS-1))
# COMPARE WITH TOLERABLE MEAN SQUARE DIFFERENCE
TL=0.0
TH=1.E+08
TD=1.
CALL RNGETR('THRESHOLD MSV?.',TL,TH,TD,THRESH,IEV,%9000)
IF(SUM<=THRESH) GOTO 5000
# SOFT CORRECTION
SOFCOR=SQRT(THRESH/SUM)
SOFCOR=1.-SOFCOR
WRITE(TTYOT,1005) SOFCOR
1005 FORMAT(' SOFCOR= ',E9.4)
#
WRITE(TTYOT,1010) THRESH, SUM
1010 FORMAT(' VARIANCE THRESHOLD= ',E9.4,' MS DIFF= ',E9.4)
DO BLKN=1,YS
$(
CALL RREAD(FDI2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
CALL RREAD(FDI2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
DO I=1,XS
$(
ILIN1(I)=ILIN1(I)+SOFCOR*CORR(BLKN,I)
ILIN2(I)=ILIN2(I)+SOFCOR*CORI(BLKN,I)
$)
$)
IF(XS>=1)
$(
DO I=NS,NPPL
$(
ILIN1(I)=ILIN1(I)+SOFCOR*CORR(BLKN,I)
ILIN2(I)=ILIN2(I)+SOFCOR*CORI(BLKN,I)
$)
$)
CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
CALL RWRITE(FD01,ILIN2,2,BLKN,JDENT,.WAIT,IEV,%9000)
$)
IF(YS>=1)
$(
DO BLKN=MS,NLIN
$(

```



```

CALL RREAD(FDI2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
CALL RREAD(FDI2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
DO I=1,XS
$(
  ILIN1(I)=ILIN1(I)+SOF COR*CORR(BLKN,I)
  ILIN2(I)=ILIN2(I)+SOF COR*CORI(BLKN,I)
$)
IF(XS>=1)
$(
  DO I=NS,NPPL
  $(
    ILIN1(I)=ILIN1(I)+SOF COR*CORR(BLKN,I)
    ILIN2(I)=ILIN2(I)+SOF COR*CORI(BLKN,I)
  $)
$)
CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
CALL RWRITE(FD01,ILIN2,2,BLKN,JDENT,.WAIT,IEV,%9000)
$)
#
# COPY OUTSIDE MEASUREMENT DOMAIN
MSM1=MS-1
YSP1=YS+1
IF(MSM1=YSP1)
$(
  DO BLKN=YSP1,MSM1
  $(
    CALL RREAD(FDI2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
    CALL RREAD(FDI2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
    CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
    CALL RWRITE(FD01,ILIN2,2,BLKN,JDENT,.WAIT,IEV,%9000)
  $)
$)
GOTO 6000
#
# NO CORRECTION, STRAIGHT COPY
5000 CONTINUE
WRITE(TTYOT,1020) THRESH
1020 FORMAT(' THRESHOLD= ',E8.2,' SATISFIED')
DO BLKN=1,NLIN
$(
  CALL RREAD(FDI2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
  CALL RREAD(FDI2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
  CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
  CALL RWRITE(FD01,ILIN2,2,BLKN,JDENT,.WAIT,IEV,%9000)
$)
#
# 6000 CONTINUE
#
CALL PPOP
CALL CLOSE(FDI1)
CALL CLOSE(FDI2)
CALL CLOSE(FD01)
RETURN
#
# ERROR CONDITIONS
9000 CONTINUE
CALL CLOSE(FDI1)
CALL CLOSE(FDI2)
CALL CLOSE(FD01)
#

```

9999 CONTINUE  
9090 RETURN 1  
END

```

#--MSMTSD                                DRIVER
#
#IDENTIFICATION BEEX, RATFOR, VAX-11
#
#PURPOSE
#       DRIVER FOR COMMAND THAT GENERATES FREQUENCY LIMITED MSMTS
#       OVER THE MSMT DOMAIN
#
#ENTRY POINT MSMTSD(WORK,ERRET)
#
#*****
#
#INCLUDE MACAL
#SUBROUTINE MSMTSD(WORK,*)
#IMPLICIT INTEGER (A - Z)
#INCLUDE GIPCOM
#INCLUDE ERROR
#INTEGER IDENT(.IDLENGTH)
#DIMENSION WORK(.ARB)
#
#EQUIVALENCE (NPPL,IDENT(.IDNPPL))
#EQUIVALENCE (NLIN,IDENT(.IDNLINS))
#EQUIVALENCE (MODE,IDENT(.IDMODE))
#
#CALL PPUSH('MSMTSD')
#                                OPEN INPUTFILE
#CALL RDKINL(FDI1,IDENT,.OLD,IEV,%9999)
#CALL CLOSE(FDI1)
#                                CHECK INPUTFILE
#IF(MODE^=.REALMODE) GOTO 9000
#
#NXT=1
#ILIN1=GETWP(NXT,.REALMODE,NPPL)
#ILIN2=GETWP(NXT,.REALMODE,NPPL)
#
#IF(.OK^=OSALOC(NXT)) GOTO 9010
#
#CALL COMTIN(%9999)
#
#CALL MSMTSA(FDI1,FDO1,FDO2,WORK(ILIN1),WORK(ILIN2),
#           NLIN,NPPL,IEV,%9999)
#
#CALL PPOP
#
#RETURN
#
#                                ERROR CONDITIONS
#9000 CONTINUE
#                                ILLEGAL DATAMODE
#IEV=-2012
#GOTO 9999
#
#9010 CONTINUE
#IEV=OSGIEV(IEV)
#GOTO 9999
#
#9999 CONTINUE
#CALL CLOSE(FDI1)
#CALL CLOSE(FDO1)
#CALL CLOSE(FDO2)
#
#RETURN 1
#END

```

```

# MSMTSA                                ALGORITHM
#
# IDENTIFICATION BEEH, RATFOR, VAX-11
#
# PURPOSE
#     ALGORITHM SECTION FOR COMMAND THAT GENERATES
#     FREQUENCY LIMITED MSMTS OVER THE MSMT DOMAIN
#
# ENTRY POINT MSMTSA(FDI,FDO1,FDO2,ILIN1,ILIN2,NLIN,NPPL,IEV,*)
#
# *****
#
# INCLUDE MACAL
# SUBROUTINE MSMTSA(FDI,FDO1,FDO2,ILIN1,ILIN2,NLIN,NPPL,IEV,*)
# IMPLICIT INTEGER (A - Z)
# CHARACTER FDI1(.FDLENGTH), FDO1(.FDLENGTH), FDO2(.FDLENGTH)
# INTEGER IDENT(.IDLENGTH), JDENT(.IDLENGTH), LDENT(.IDLENGTH)
# REAL ILIN1(NPPL), ILIN2(NPPL), MSV, CONST
# INTEGER*4 DUMMY
#
# INCLUDE TCOM
#
# CALL PPUSH('MSMTSA')
#
#     OPEN INPUTFILE
# CALL CPYIDR(FDI1,IDENT,.OPNTMP,IEV,%9000)
# CALL DSCNAM('MSMTSA',IEV,%9000)
#
# DO I=6,19
# $(
# IF(I .EQ. 6 .OR. I .EQ. 7 .OR. I .EQ. 13 .OR. I .EQ. 14 .OR.
#   I .EQ. 17 .OR. I .EQ. 19) $(
#
#     X=IDENT(I)
#     JDENT(I)=X
#     LDENT(I)=X
#
# $)
#
#     OPEN OUTPUTFILE
# CALL COPYDS(FDO1,JDENT,IEV,%9000)
#
# YSL=1
# YSH=NLIN/2-1
# YSD=(YSL+YSH)/2
# CALL RNGETI('VERTICAL SIZE MSMT
# DOMAIN?.',YSL,YSH,YSD,YS,IEV,%9000)  XSL=1
# XSH=NPPL/2-1
# XSD=(XSL+XSH)/2
# CALL RNGETI('HORIZONTAL SIZE MSMT
# DOMAIN?.',XSL,XSH,XSD,XS,IEV,%9000)  NOIVL=0
# NOIVH=100000
# NOIVD=0
# CALL RNGETI('NOISE VARIANCE?.',NOIVL,NOIVH,NOIVD,NOIV,IEV,%9000)
#
# CONST=SQRT(12.*NOIV)
# IF(NOIV==NOIVD) CONST=0.0
# DUMMY=34589
# DO BLKN=1,NLIN
# $(
#   CALL RREAD(FDI1,ILIN1,1,BLKN,IDENT,.WAIT,IEV,%9000)
#   DO I=1,NPPL
#   $(
#     ILIN1(I)=ILIN1(I)+CONST*(RAN(DUMMY)-.5)
#   $)
# $)

```

```

CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
CALL RREAD(FD11,ILIN1,2,BLKN,IDENT,.WAIT,IEV,%9000)
DO I=1,NPPL
$(
  ILIN1(I)=ILIN1(I)+CONST*(RAN(DUMMY)-.5)
$)
CALL RWRITE(FD01,ILIN1,2,BLKN,JDENT,.WAIT,IEV,%9000)
$)
#
BLKN=NLIN/2+1
DO I=1,NPPL
$(
  ILIN1(I)=0.
$)
ILIN1(1)=YS+.02
ILIN1(2)=XS+.02
CALL RWRITE(FD01,ILIN1,1,BLKN,JDENT,.WAIT,IEV,%9000)
#
CALL CLOSE(FD01)
CALL CLOSE(FD11)
CALL CPYIDR(FD11,IDENT,.OPNTMP,IEV,%9000)
CALL COPYDS(FD02,LDENT,IEV,%9000)
XSP1=XS+1
NS=NPPL-XS+2
NSM1=NS-1
YSP1=YS+1
MS=NLIN-YS+2
MSM1=MS-1
#
MSV=0.0
DO BLKN=1,YS
$(
  CALL RREAD(FD11,ILIN1,1,BLKN,IDENT,.WAIT,IEV,%9000)
  CALL RREAD(FD11,ILIN2,2,BLKN,IDENT,.WAIT,IEV,%9000)
  DO I=1,XS
  $(
    MSV=MSV+ILIN1(I)*ILIN1(I)+ILIN2(I)*ILIN2(I)
  $)
  DO I=NS,NPPL
  $(
    MSV=MSV+ILIN1(I)*ILIN1(I)+ILIN2(I)*ILIN2(I)
  $)
  DO I=XSP1,NSM1
  $(
    ILIN1(I)=0.0
    ILIN2(I)=0.0
  $)
  CALL RWRITE(FD02,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
  CALL RWRITE(FD02,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
$)
DO BLKN=MS,NLIN
$(
  CALL RREAD(FD11,ILIN1,1,BLKN,IDENT,.WAIT,IEV,%9000)
  CALL RREAD(FD11,ILIN2,2,BLKN,IDENT,.WAIT,IEV,%9000)
  DO I=1,XS
  $(
    MSV=MSV+ILIN1(I)*ILIN1(I)+ILIN2(I)*ILIN2(I)
  $)

```

```

DO I=NS,NPPL
$(
  MSV=MSV+ILIN1(I)*ILIN1(I)+ILIN2(I)*ILIN2(I)
$)
DO I=XSP1,NSM1
$(
  ILIN1(I)=0.0
  ILIN2(I)=0.0
$)
CALL RWRITE(FDO2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
CALL RWRITE(FDO2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
$)
DO BLKN=YSP1,MSM1
$(
  DO I=1,NPPL
  $(
    ILIN1(I)=0.0
    ILIN2(I)=0.0
  $)
  CALL RWRITE(FDO2,ILIN1,1,BLKN,LDENT,.WAIT,IEV,%9000)
  CALL RWRITE(FDO2,ILIN2,2,BLKN,LDENT,.WAIT,IEV,%9000)
$)
MSV=MSV/((2*XS-1)*(2*YS-1))
WRITE(TTYOT,1010) MSV
1010 FORMAT(' MEAN SQUARE VALUE OF MSMTS= ',E10.4)
#
CALL PPOP
CALL CLOSE(FDI1)
CALL CLOSE(FDO1)
CALL CLOSE(FDO2)
RETURN
#
ERROR CONDITIONS
9000 CONTINUE
CALL CLOSE(FDI1)
CALL CLOSE(FDO1)
CALL CLOSE(FDO2)
#
9999 CONTINUE
9090 RETURN 1
END

```

INFORMATION TRANSFERS PUBLISHED, PRESENTED, AND IN PREPARATION

"2-D Spectral Estimator Approaches to Enhanced Scene Resolution,"  
presented by A.A. (Louis) Beex at the 1982 Virginia IEEE Conference  
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"Iterative Reconstruction of Space-Limited Scenes from Noisy Frequency-  
Limited Measurements," presented and published by A.A. (Louis) Beex  
at the IEEE International Conference on Acoustics, Speech, and Signal  
Processing (ICASSP '83), Boston, MA. pp. 147 - 150, April 14 - 16, 1983.

"Soft Constraint Iterative Reconstruction from Noisy Projections,"  
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